



# *Introduction to Probability and Statistics*

## *Chapter 5*

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## Chapter 5

# Joint Probability Distributions and Random Samples

# Chapter Outlines

**5.4 The Distribution of the Sample Mean**

**5.5 The Distribution of a Linear Combination**

## 5.4 The Distribution of the Sample Mean

### Using the Sample Mean

Let  $X_1, \dots, X_n$  be a random sample from a distribution with mean value  $\mu$  and standard deviation  $\sigma$ . Then

$$1. E(\bar{X}) = \mu_{\bar{X}} = \mu$$

$$2. V(\bar{X}) = \sigma_{\bar{X}}^2 = \frac{\sigma^2}{n} \text{ and } \sigma_{\bar{X}} = \frac{\sigma}{\sqrt{n}}$$

In addition, with  $T_0 = X_1 + \dots + X_n$  (The sample total)

$$E(T_0) = n\mu \quad V(T_0) = n\sigma^2 \quad \sigma_{T_0} = \sqrt{n} \sigma$$

### Example 5.24 (p. 213)

Let  $X_1, \dots, X_{25}$  be a random sample from a distribution with mean  $\mu=28000$  and standard deviation  $\sigma=5000$ . Find:

- 1)  $E(\bar{X})$
- 2)  $V(\bar{X})$
- 3)  $E(T)$  and
- 4)  $V(T)$ , where  $T = X_1 + \dots + X_{25}$

## Normal Population Distribution

Let  $X_1, \dots, X_n$  be a random sample from a normal distribution with mean value  $\mu$  and standard deviation  $\sigma$ . Then for any  $n$ :

$$1) \quad \bar{X} \sim N(\mu, \sigma^2 / n)$$

$$2) \quad T \sim N(n\mu, n\sigma^2)$$

### Example 5.25 (p. 214)

The time that it takes a randomly selected rat of a certain subspecies to find its way through a maze is a normally distributed rv with  $\mu = 1.5$  min and  $\sigma = 0.35$  min. Suppose 5 rats are selected. **Find:**

- 1) The probability that the total time for the 5 rats is between 6 and 8 min?
- 2) The probability that the average time for the 5 rats is at most 2 min?

Solution:

$$1) \quad \Phi(0.64) - \Phi(-1.92) = .7115$$

$$2) \quad \Phi(3.191) = .9993$$

# The Central Limit Theorem

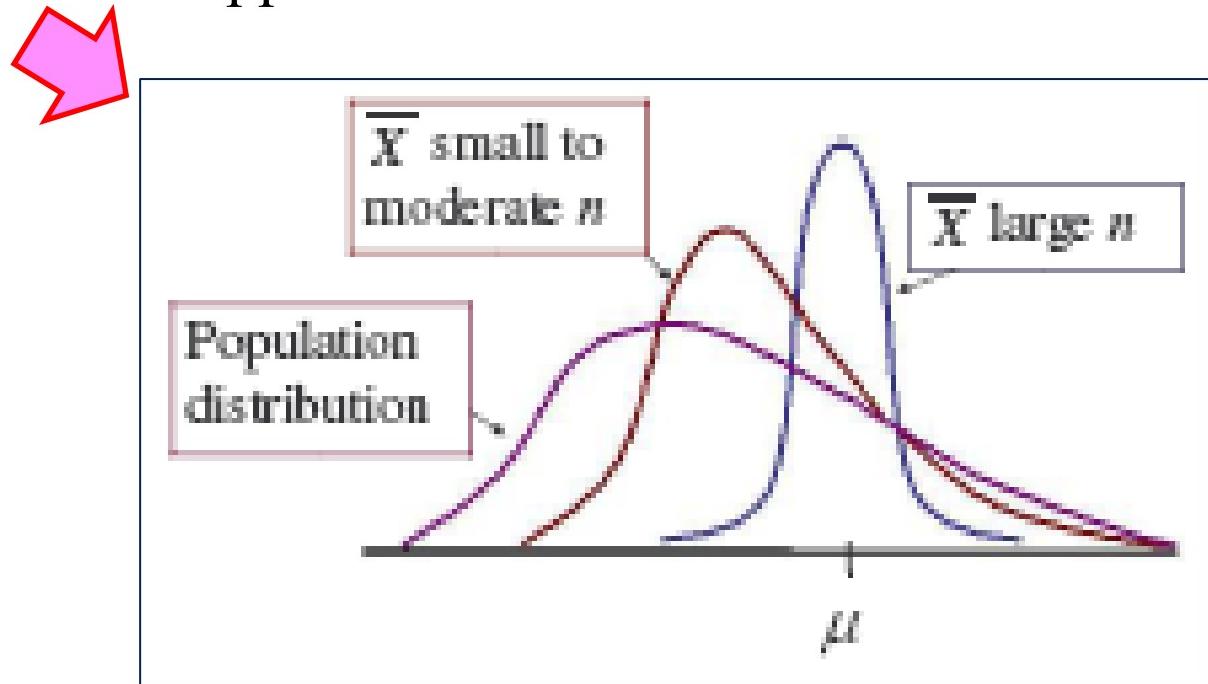
Let  $X_1, \dots, X_n$  be a random sample from a distribution with mean value  $\mu$  and standard deviation  $\sigma$ . Then if  $n$  is sufficiently large,

- 1)  $\bar{X} \sim \text{approximately } N(\mu, \sigma^2 / n)$
- 2)  $T \sim \text{approximately } N(n\mu, n\sigma^2)$

The larger value of  $n$ , the better approximation.

Rule of Thumb

If  $n > 30$ , the Central Limit Theorem can be used.



## Example 5.26 (p. 215)

The amount of a particular impurity in a batch of a certain chemical product is a random variable with mean value 4.0 g and standard deviation 1.5 g. If 50 batches are independently prepared, what is the (approximate) probability that

- 1) the sample average amount of impurity is between 3.5 and 3.8 g?
- 2) the sample total amount of impurity is between 175 and 190 g?

Solution:

$$1) \quad n = 50 > 30 \Rightarrow \bar{X} \sim \text{approximately } N(4, .2121^2)$$

$$P(3.5 \leq \bar{X} \leq 3.8) \approx \left( \frac{3.5 - 4}{.2121} \leq Z \leq \frac{3.8 - 4}{.2121} \right)$$

$$= \Phi(-0.94) - \Phi(-2.36) = .1645$$

$$2) \quad T \sim \text{approximately } N(50 \times 4, 50 \times 1.5^2)$$

## 5.5 The Distribution of a Linear Combination

### Linear Combination

Given a collection of  $n$  random variables  $X_1, \dots, X_n$  and  $n$  numerical constants  $a_1, \dots, a_n$ , the rv

$$Y = a_1 X_1 + \dots + a_n X_n = \sum_{i=1}^n a_i X_i$$

is called a *linear combination* of the  $X_i$ 's.

Examples:

- 1) Taking  $a_i = 1$ ,  $i=1,2,\dots,n$ , gives  $Y = X_1 + \dots + X_n$  (*Total sum*)
- 2) Taking  $a_i = 1/n$ ,  $i=1,2,\dots,n$ , gives  $Y = (X_1 + \dots + X_n)/n$  (*Sample mean*)

## Expected Value of a Linear Combination

Let  $X_1, \dots, X_n$  have mean values  $\mu_1, \mu_2, \dots, \mu_n$  and variances of  $\sigma_1^2, \sigma_2^2, \dots, \sigma_n^2$  respectively. Whether or not the  $X_i$ 's are independent

$$E(a_1 X_1 + \dots + a_n X_n) = \sum_{i=1}^n a_i E(X_i) = \sum_{i=1}^n a_i \mu_i$$

### Variance of a Linear Combination

If  $X_1, \dots, X_n$  are independent

$$V(a_1 X_1 + \dots + a_n X_n) = \sum_{i=1}^n a_i^2 V(X_i) = \sum_{i=1}^n a_i^2 \sigma_i^2$$

and

$$\sigma_{a_1 X_1 + \dots + a_n X_n} = \sqrt{a_1^2 \sigma_1^2 + \dots + a_n^2 \sigma_n^2}$$

## Examples:

- 1)  $E(2 X_1 - 3 X_2) = 2 E(X_1) - 3 E(X_2)$
- 2)  $V(2 X_1 - 3 X_2) = 2^2 V(X_1) + (-3)^2 V(X_2)$
- 3)  $\sigma_{2X_1-3X_2} = \sqrt{2^2 \sigma_1^2 + 3^2 \sigma_2^2}$
- 4) A gas station sells three grades of gasoline: regular, extra, and super. These are priced at \$21.2, \$21.35, and \$21.5 per gallon, respectively. Let  $X_1$ ,  $X_2$  and  $X_3$  denote the amount of these grades purchased (gallon) on a particular day. Suppose  $X_i$ 's independent with  $\mu_1=1000$ ,  $\mu_2=500$ ,  $\mu_3=300$  and  $\sigma_1=100$ ,  $\sigma_2=80$ ,  $\sigma_3=50$ . Find
  - a) The expected revenue from sales on that day;
  - b) The variance of the revenue from sales;
  - c) The standard deviation of the revenue from sales;

Solution:

Let  $Y$  be the revenue from sales. Then  $Y = 21.2 X_1 + 21.35 X_2 + 21.5 X_3$

a)  $E(Y) = \$4,125$       b)  $V(Y) = 104,025$       c)  $\sigma_Y = \$322.529$

## Variance of a Linear Combination

For any  $X_1, \dots, X_n$ ,

$$V(a_1 X_1 + \dots + a_n X_n) = \sum_{i=1}^n \sum_{j=1}^n a_i a_j Cov(X_i, X_j)$$

where

$$Cov(X_i, X_j) = E[(X_i - \mu_i)(X_j - \mu_j)]$$

## Difference Between Two Random Variables

$$E( X_1 - X_2 ) = E( X_1 ) - E( X_2 )$$

and, if  $X_1$  and  $X_2$  are independent

$$V( X_1 - X_2 ) = V( X_1 ) + V( X_2 )$$

**Example:** A certain automobile manufacture equips a particular model with either a six-cylinder engine or a four-cylinder engine. Let  $X_1$  and  $X_2$  be fuel efficiencies for independently and randomly selected six-cylinder and four-cylinder cars, respectively. With  $\mu_1=22$ ,  $\mu_2=26$ ,  $\sigma_1=1.2$  and  $\sigma_2=1.5$ . Find

- a)  $E(X_1 - X_2)$
- b)  $V(X_1 - X_2)$
- c)  $\sigma_{X_1 - X_2}$

## Difference Between Normal Random Variables

If  $X_1, X_2, \dots, X_n$  are independent, *normally* distributed rv's, then any linear combination of the  $X_i$ 's also has a normal distribution. The difference  $X_1 - X_2$  between two independent, normally distributed variables is itself normally distributed.

**Example:** In example 4, slide 10, if  $X_i$ 's are normally distributed find:

- (1) probability that revenue exceeds 4500?
- (2) Probability that the amount of regular gas purchased exceeds that of supper by 500 gallon?

**Solution:**

$$\begin{aligned}(1) \quad Y &\sim N(4,125, 104,025), \text{ then } P(Y > 4500) = P\left(Z > \frac{4500 - 4125}{322.53}\right) \\ &= P(Z > 1.16) = 1 - \Phi(1.16) = 0.123\end{aligned}$$

- (2) In class.